

# Linear Algebra

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# The solution of Linear Equations

$$\left\{ \begin{array}{ccccccc} a_{11}x_1 & + & \cdots & + & a_{1n}x_n & = & b_1 \\ \vdots & & \vdots & & \vdots & = & \vdots \\ a_{n1}x_1 & + & \cdots & + & a_{nn}x_n & = & b_n \end{array} \right.$$

$n$  equation in  $n$  unknowns.

- Matrix form
- Row picture
- Column picture

## Example for $n = 2$

$$\begin{cases} x_1 - x_2 = 1 \\ x_1 + 2x_2 = 4 \end{cases}$$

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$Ax = b$

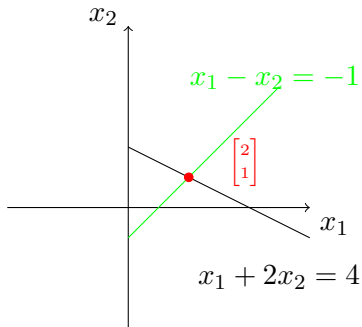
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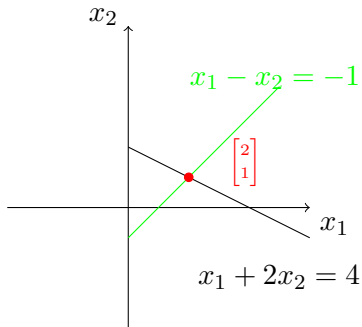
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$Ax = b$

- Column picture

$$x_1 \underbrace{\begin{bmatrix} 1 \\ 1 \end{bmatrix}} + x_2 \underbrace{\begin{bmatrix} -1 \\ 2 \end{bmatrix}} = \begin{bmatrix} 1 \\ 4 \end{bmatrix}$$

a linear combination of  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$

## Example for $n = 3$

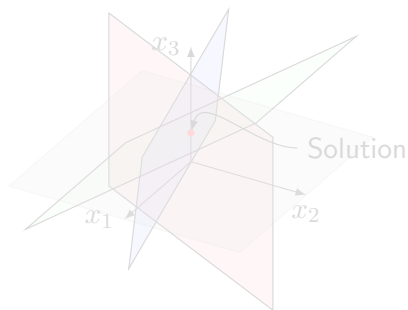
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$$\begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & 3 & 4 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \\ 4 \end{bmatrix}$$

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• Row picture



• Column picture

$$x_1 \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ 2 \\ 3 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ -1 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \\ 4 \end{bmatrix}$$

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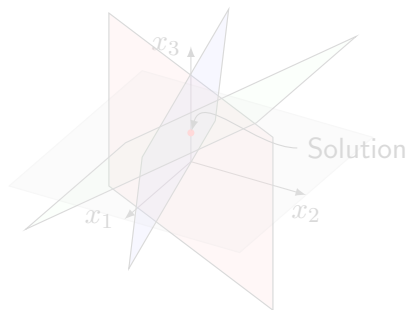
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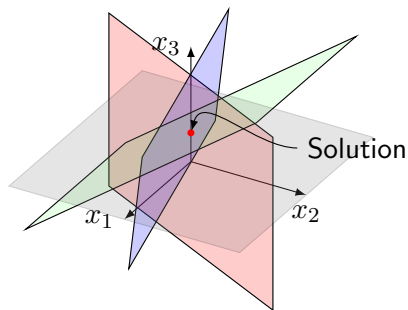
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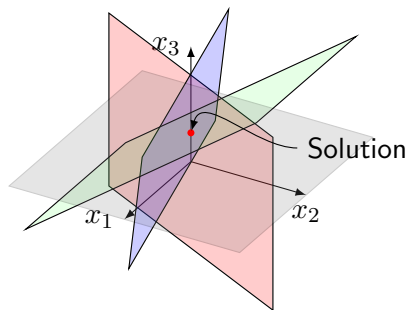
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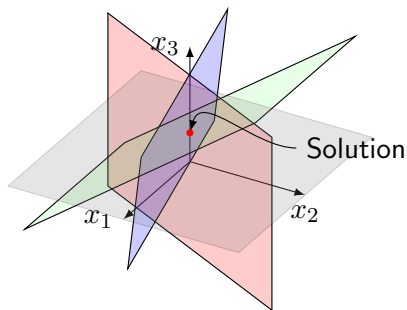
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For another given  $b \in \mathbb{R}^3$

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The equation system  $Ax = b$  has a solution.

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- Do all linear combinations of  $\begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} -1 \\ 2 \\ 3 \end{bmatrix}$  and  $\begin{bmatrix} 0 \\ -1 \\ 4 \end{bmatrix}$  fill  $\mathbb{R}^3$ ?



# Linear Independence

$$A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & 3 & 4 \end{bmatrix}$$

$$A = \begin{bmatrix} 2 & 4 & 0 \\ -1 & -2 & -1 \\ 0 & 0 & 4 \end{bmatrix}$$

$$A = \begin{bmatrix} 2 & 1 & 3 \\ -1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

# Linear function picture of the linear equation system

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Two following questions are equivalent:

Can  $Ax = b$  be solved for a given  $b \in \mathbb{R}^3$ ?

$b \in \text{range}(T)$ ?

The properties of

$$T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$T(x) = Ax$$

- ①  $T(u + v) = T(u) + T(v)$  for all  $u, v \in \mathbb{R}^3$ .
- ②  $T(cu) = cT(u)$  for all  $c \in \mathbb{R}$ .

Every function from  $\mathbb{R}^n$  to  $\mathbb{R}^n$  with two above properties is called linear function.

# An Example of a linear function and its representation

$$T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$$
$$T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 + x_2 \\ x_2 + x_3 \\ 0 \end{bmatrix}$$

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$$\begin{aligned} T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) &= xT \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) + yT \left( \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right) + zT \left( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right) \\ xT \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right) + yT \left( \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \right) + zT \left( \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right) &= \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \end{aligned}$$



# Matrix Representations of Linear Function

Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a linear function and  $u = \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix} \in \mathbb{R}^n$ .

In the  $n$ -dimensional Euclidean space  $\mathbb{R}^n$ , the standard basis consists of  $n$  distinct vectors

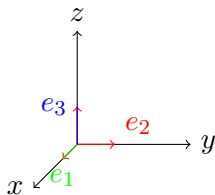
$$E = \{e_i \mid 1 \leq i \leq n\}$$

where  $e_i$  denotes the vector with a 1 in the  $i$ -th coordinate and 0's elsewhere. Since  $T$  is a linear function, we have

$$\begin{aligned} T(u) &= T(u_1 e_1 + \dots + u_n e_n) = u_1 T(e_1) + \dots + u_n T(e_n) \\ &= \begin{bmatrix} T(e_1) & \dots & T(e_n) \end{bmatrix} \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix} \end{aligned}$$

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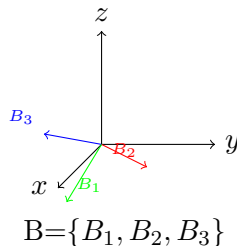
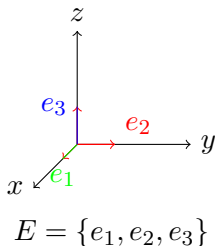
The matrix  $\begin{bmatrix} T(e_1) & \cdots & T(e_n) \end{bmatrix}$  is called the matrix representation of linear function (transformation)  $T$  which is denoted by  $[T]_E$ .



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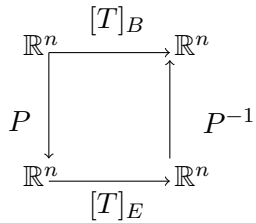
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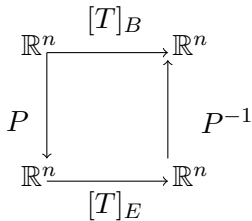


What is the relation between  $[T]_B$  and  $[T]_E$ ?

# Change of basis



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$$[T]_B = P^{-1}[T]_E P$$

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- The space of all polynomials of the degree  $\leq n - 1$ .

$$P_{n-1}(x) = \{a_{n-1}x^{n-1} + \dots + a_1x + a_0 \mid a_i \in \mathbb{R} \text{ for all } 0 \leq i \leq n-1\}$$

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- Is the operation of **differentiation**

$$\frac{d}{dt} : P_3(x) \rightarrow P_2(x)$$

$$\frac{d}{dt}(a_3x^3 + a_2x^2 + a_1x + a_0) = 3a_3x^2 + 2a_2x + a_1$$

is a linear transformation?



# The matrix representation

- We have  $B = \{x^3, x^2, x, 1\}$  and  $B' = \{x^2, x, 1\}$  are bases for  $P_3(x)$  and  $P_2(x)$ , respectively.

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$$\begin{aligned} \left[ \frac{d}{dt} \right]_{\{B, B'\}} &= \begin{bmatrix} \frac{d}{dt}(x^3) & \frac{d}{dt}(x^2) & \frac{d}{dt}(x) & \frac{d}{dt}(1) \end{bmatrix} \\ &= \begin{bmatrix} 3 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \end{aligned}$$

# The operation of differentiation

$$\frac{d}{dt} : \mathbb{R}^4 \rightarrow \mathbb{R}^3$$
$$\frac{d}{dt} \begin{bmatrix} a_3 \\ a_2 \\ a_1 \\ a_0 \end{bmatrix} = \begin{bmatrix} 3 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} a_3 \\ a_2 \\ a_1 \\ a_0 \end{bmatrix} = \begin{bmatrix} 3a_3 \\ 2a_2 \\ a_1 \end{bmatrix}$$

Note. We can suppose  $\frac{d}{dt} : P_3(x) \rightarrow P_3(x)$  where  $range(\frac{d}{dt}) = P_2(x)$  and we have

$$\left[ \frac{d}{dt} \right]_B = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 3 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

A subspace  $U$  of a linear space  $V$  satisfies the following properties

- additive identity
- closed under addition
- closed under scalar multiplication

**Example.**  $U = P_2(x)$  is a subspace of  $V = P_3(x)$ .

# Invariant Subspaces

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A subspace of  $U \subset V$  is *invariant* under  $T$  if for each  $u \in U$ ,  $T(u) \in U$ .

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- Think about the Matrix representation of  $T$ !

# Eigenvalues and eigenvectors

How does a linear function  $T : V \rightarrow V$  behave on an invariant subspace of dimension 1  $U$  where

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The scalar  $\lambda$  is called an *eigenvalue* of  $T$  and the vector  $u$  is called an *eigenvector*

# Upper Triangular

Denote  $\text{span}(v_1, \dots, v_k) = \{c_1 v_1 + \dots + c_k v_k \mid c_i \in \mathbb{R} \text{ for all } 1 \leq i \leq k\}$ .

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Assume  $B = \{v_1, \dots, v_n\}$  is a basis for  $V$  such that for each  $k = 1, \dots, n$

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We have

Its matrix representation:

$$T(v_1) = a_{11}v_1$$

$$T(v_2) = a_{12}v_1 + a_{22}v_2$$

$$T(v_3) = a_{13}v_1 + a_{23}v_2 + a_{33}v_3$$

$$\vdots$$

$$T(v_n) = a_{1n}v_1 + \dots + a_{nn}v_n$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ & a_{22} & a_{23} & \cdots & a_{2n} \\ & & a_{33} & \cdots & \vdots \\ & & & \ddots & \vdots \\ & 0 & & & a_{nn} \end{bmatrix}$$



# Inner product spaces

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An inner product on  $V$  is a function  $\langle, \rangle : V \times V \rightarrow \mathbb{R}$  such that

- $\langle v, v \rangle \geq 0$  for all  $v \in V$ .
- $\langle v, v \rangle = 0$  if and only if  $v = 0$ .
- $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$  for all  $u, v, w \in V$ .
- $\langle cu, w \rangle = c \langle u, w \rangle$  for all  $u, w \in V$  and  $c \in \mathbb{R}$ .
- $\langle v, w \rangle = \langle w, v \rangle$ .

Example of an inner product on  $\mathbb{R}^n$ :

$$\left\langle \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix}, \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \right\rangle = u_1 v_1 + \dots + u_n v_n.$$

Example of an inner product on  $\mathbb{R}^n$ :

$$\left\langle \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix}, \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \right\rangle = u_1 v_1 + \dots + u_n v_n.$$

For  $v \in V$ , the *norm* of  $v$  is denoted by  $\|v\|$  and

$$\|v\| = \sqrt{\langle v, v \rangle}$$

*Thank You*